



Towards Verification and Validation for Increased Autonomy

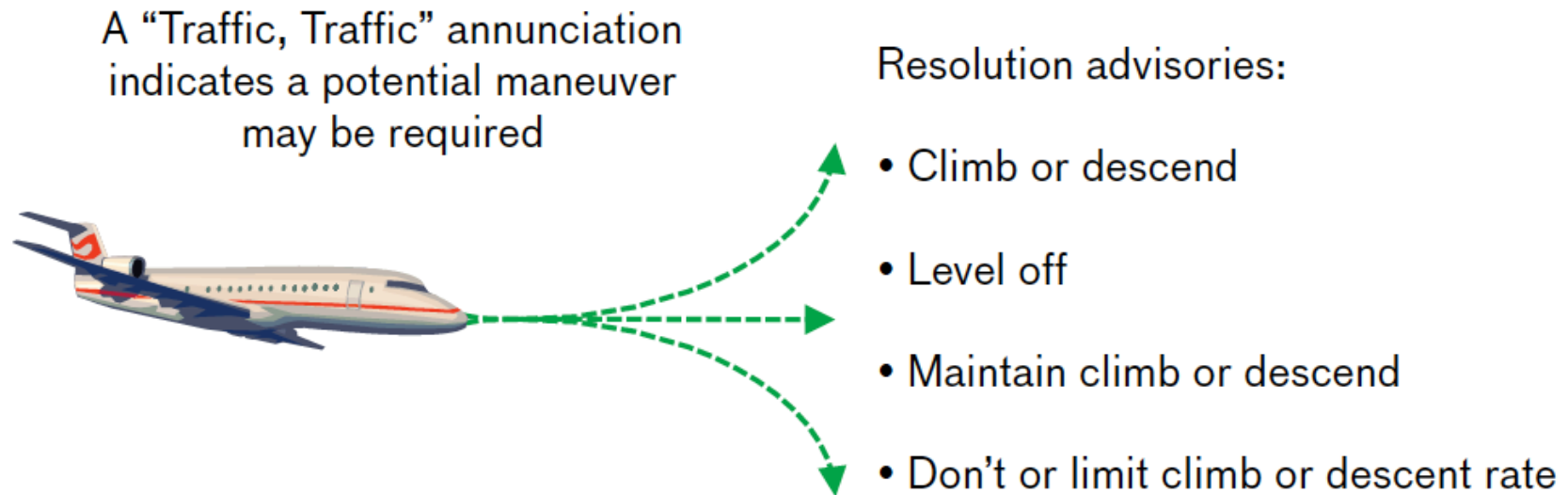
Dimitra Giannakopoulou

an aircraft is not alone in the sky...



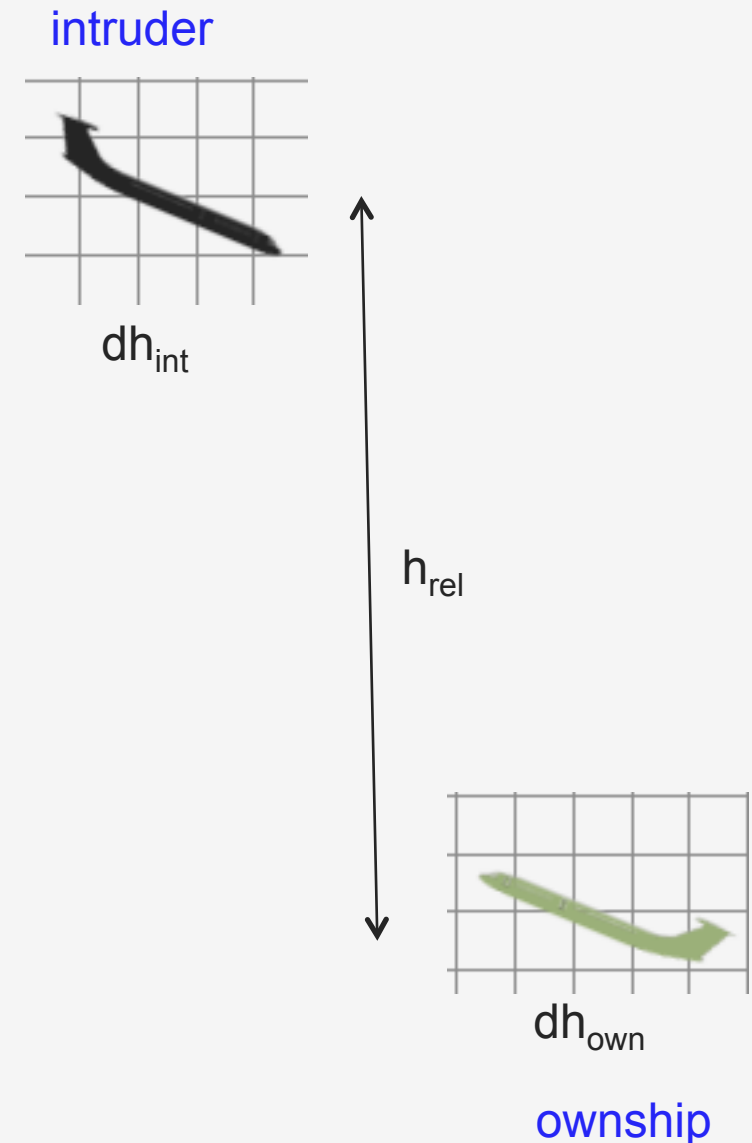
Safe Air Transportation

- Air traffic operations are expected to increase significantly. Automation must maintain or exceed current safety standards
- Separation Assurance – algorithms and systems gradually taking the role of air-traffic controllers to enable reduced aircraft separation
- Onboard-Collision Avoidance Systems – TCAS, ACAS X



ACAS X – a completely new paradigm

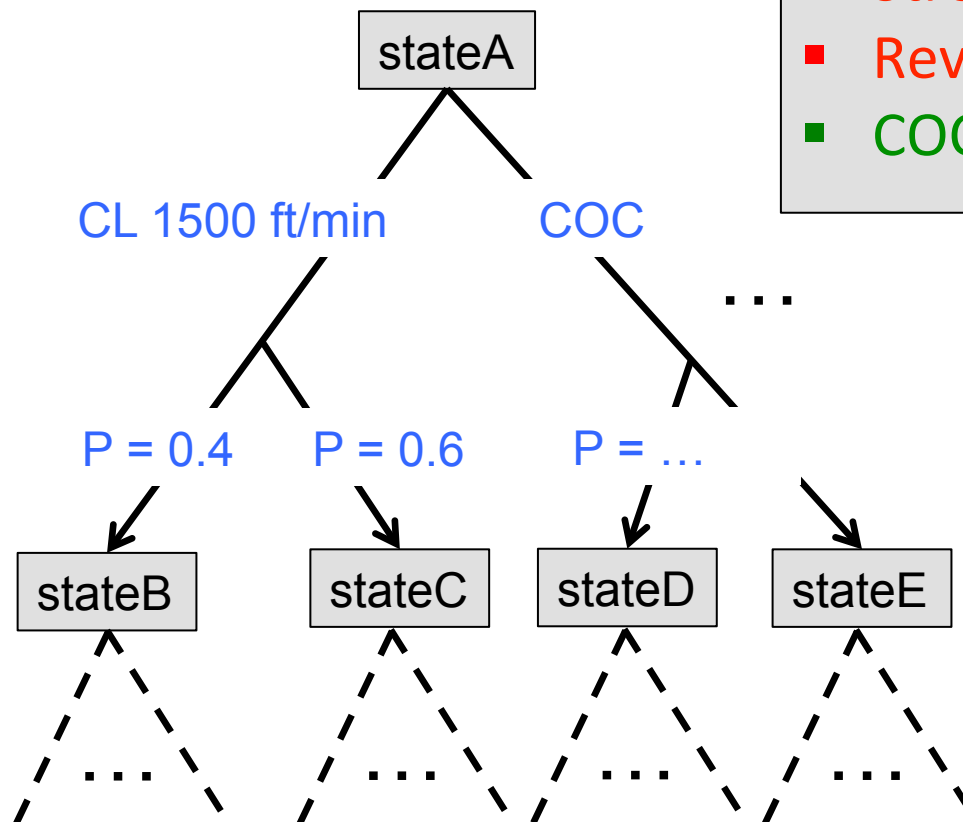
- 40 secs from Near-Mid-Air Collision (NMAC)
- state variables
 - h_{rel} : relative altitude, in $[-1000...1000]$ ft
 - dh_{own} / dh_{int} : ownship / intruder climb rates, in $[-2500...2500]$ ft/min
- a_{prev} / s_{RA} : advisory issued by ACAS X in previous sec / current pilot state, both in {COC, CL/DES1500, SCL/SDES1500, SCL/SDES2500}
- update and advisory **frequency** is set to 1 sec
- **discretization resolution** n for a variable V means that V is discretized to n points above and n points below 0 within its interval of values. For example, discretization resolution of 10 for h_{rel} means:
 $\{-1000, -900, -800, \dots, 0, 100, \dots, 900, 1000\}$



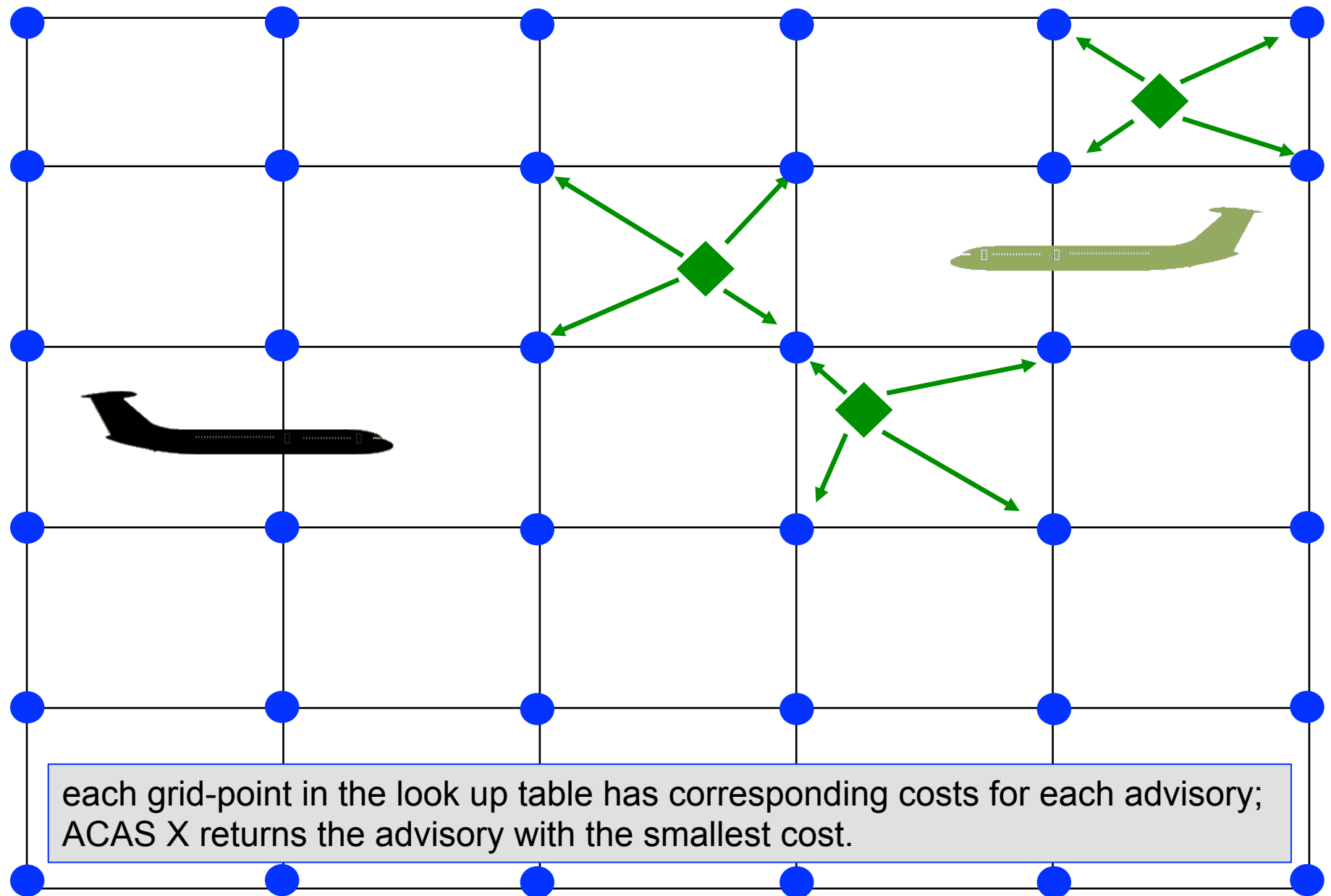
ACAS X – goals

minimize costs / maximize rewards

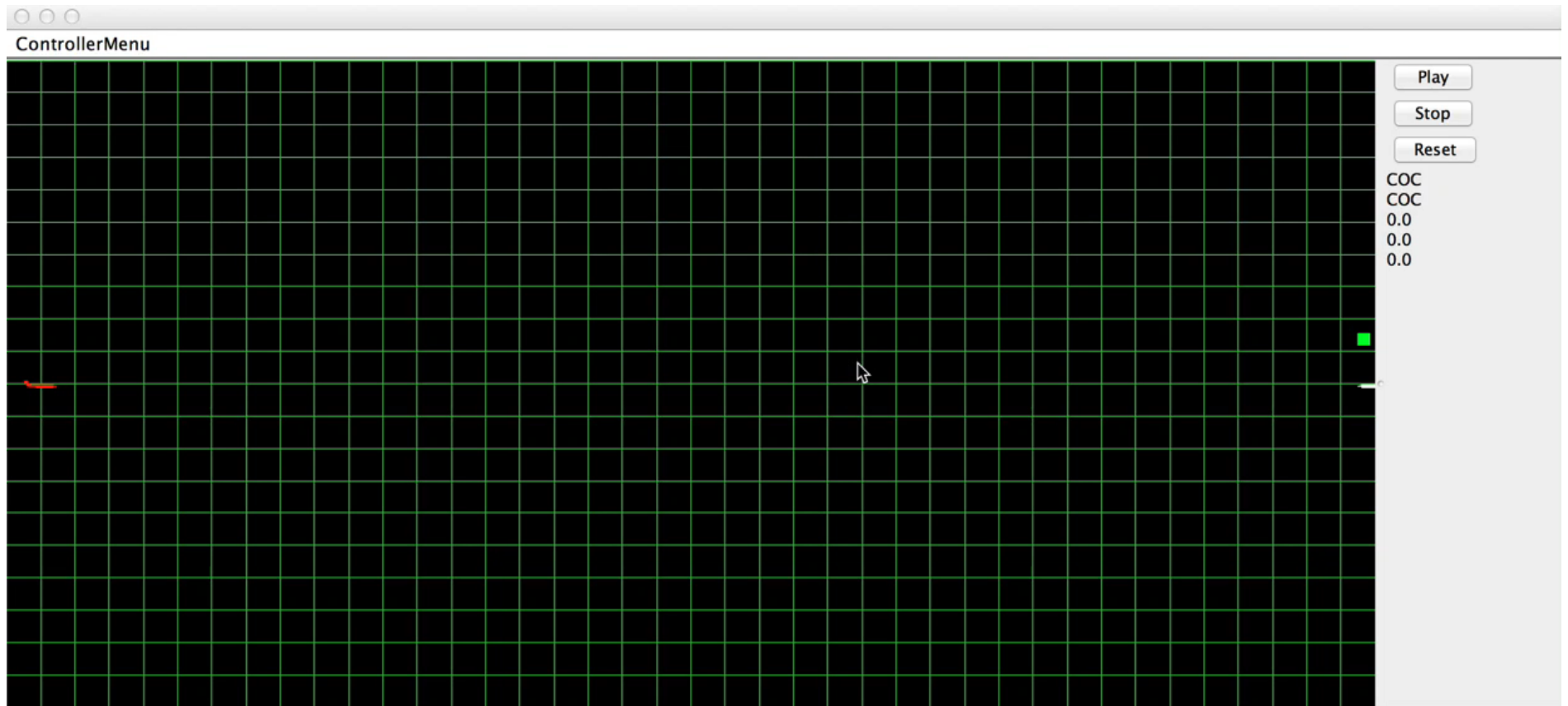
- **NMAC** (near-mid-air collision)
- **Alert** (from COC to advisory)
- **Strengthening** (strengthen advisory)
- **Reversal** (e.g. climb to descend)
- **COC** (clear of conflict)



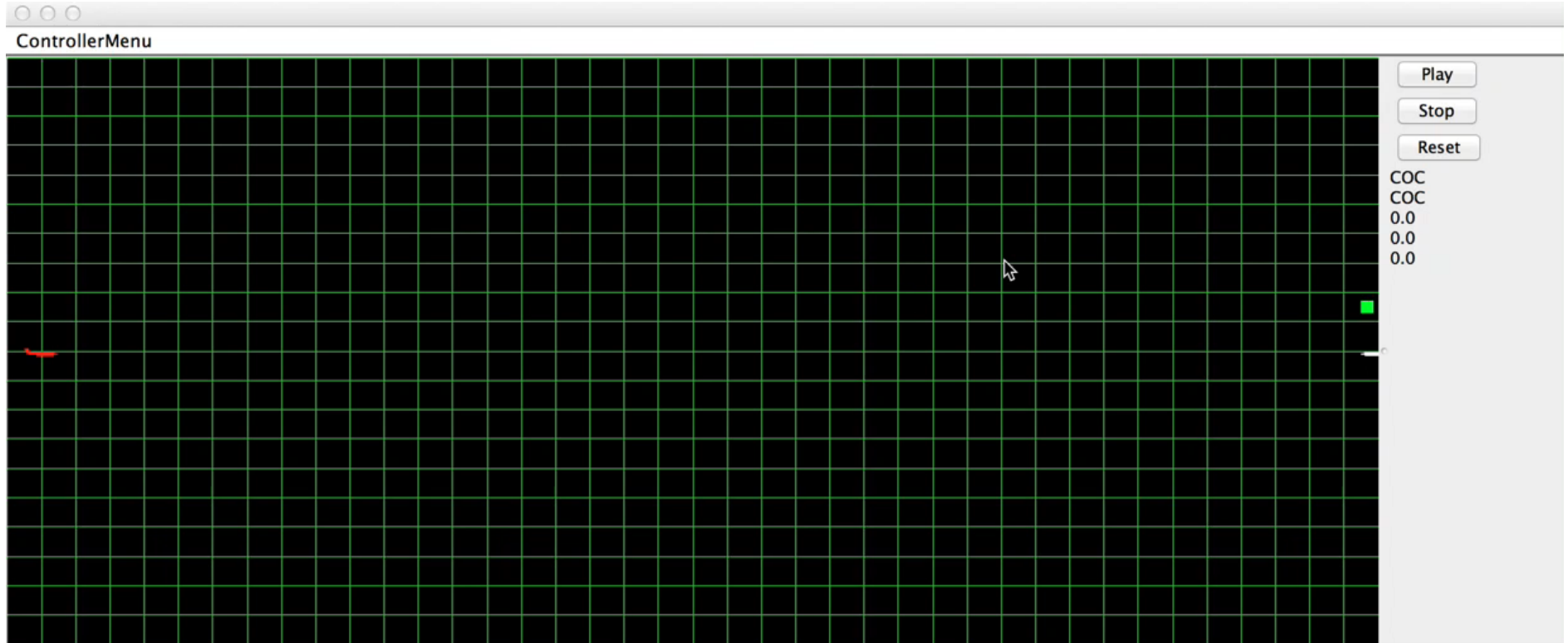
deploying ACAS X as a look up table



simulation with low NMAC weight (0.01)

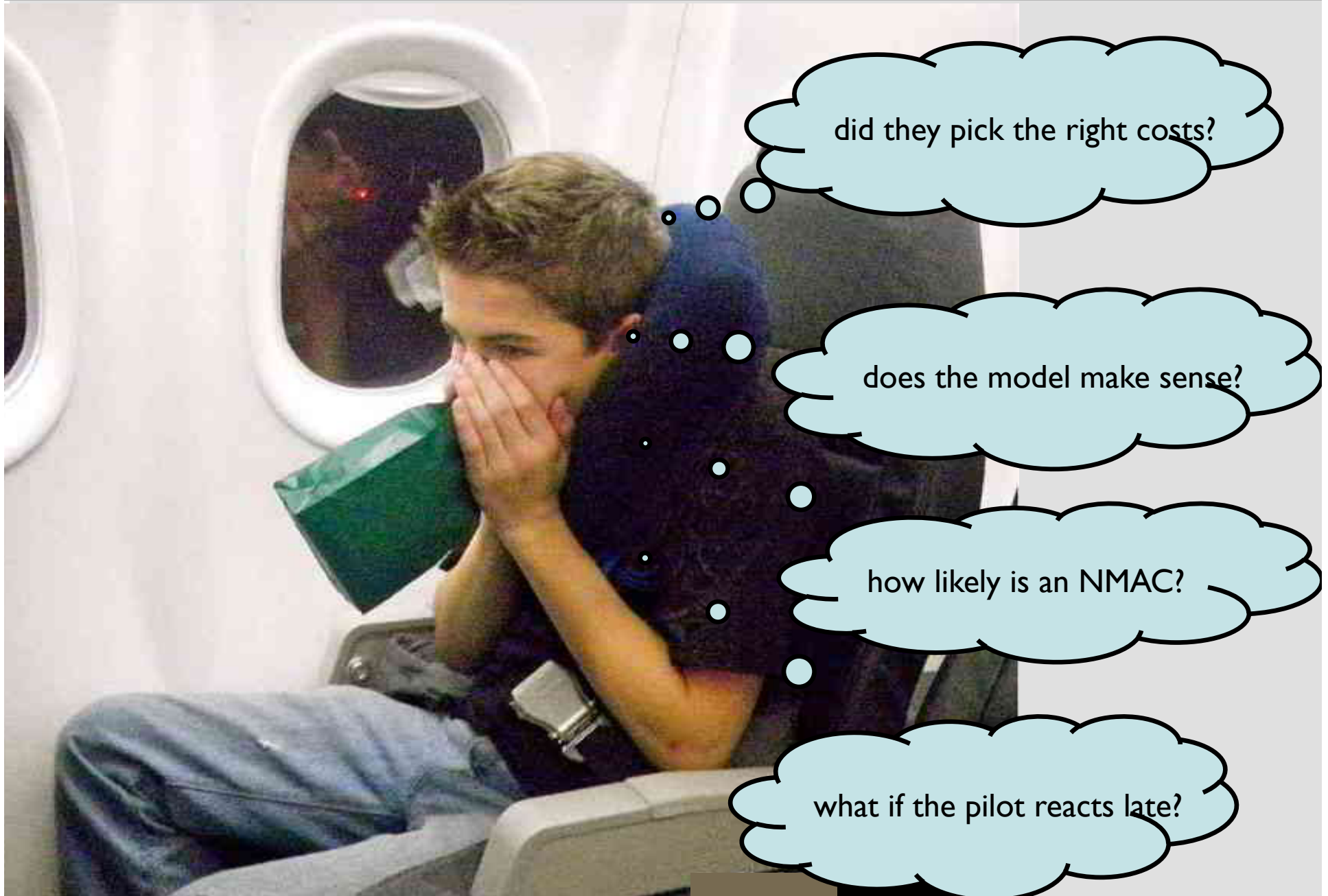


simulation with high NMAC weight (100)



verification starts with asking questions

would you trust ACAS X in a flight?



did they pick the right costs?

does the model make sense?

how likely is an NMAC?

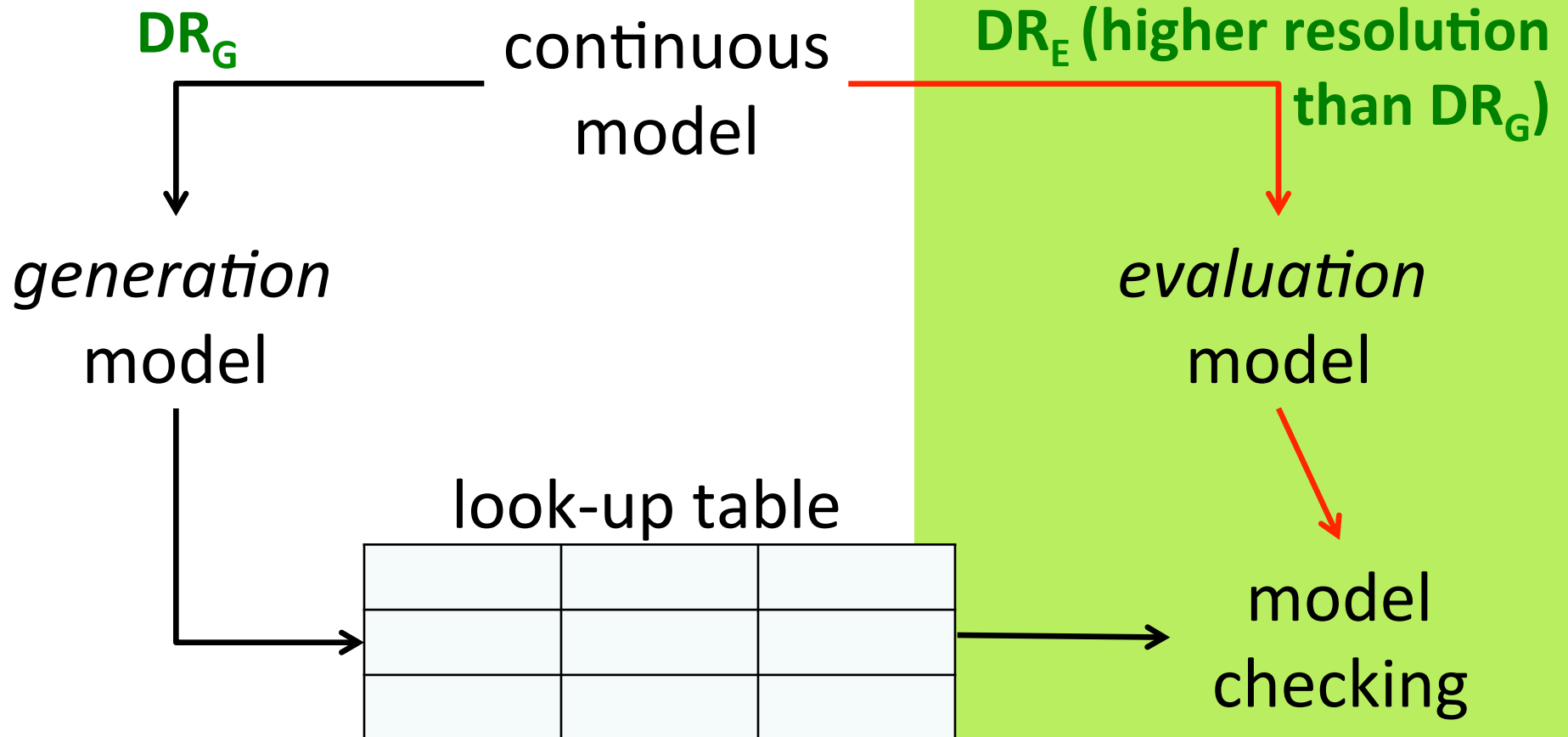
what if the pilot reacts late?

...and proceeds with answering them

discretization resolution (dh_{own} , dh_{int} , h_{rel})

DR_G : model discretization resolution for look-up table generation;
baseline [KC 2011] resolution is ($dh_{own}=10$, $dh_{int}=10$, $h_{rel}=10$)

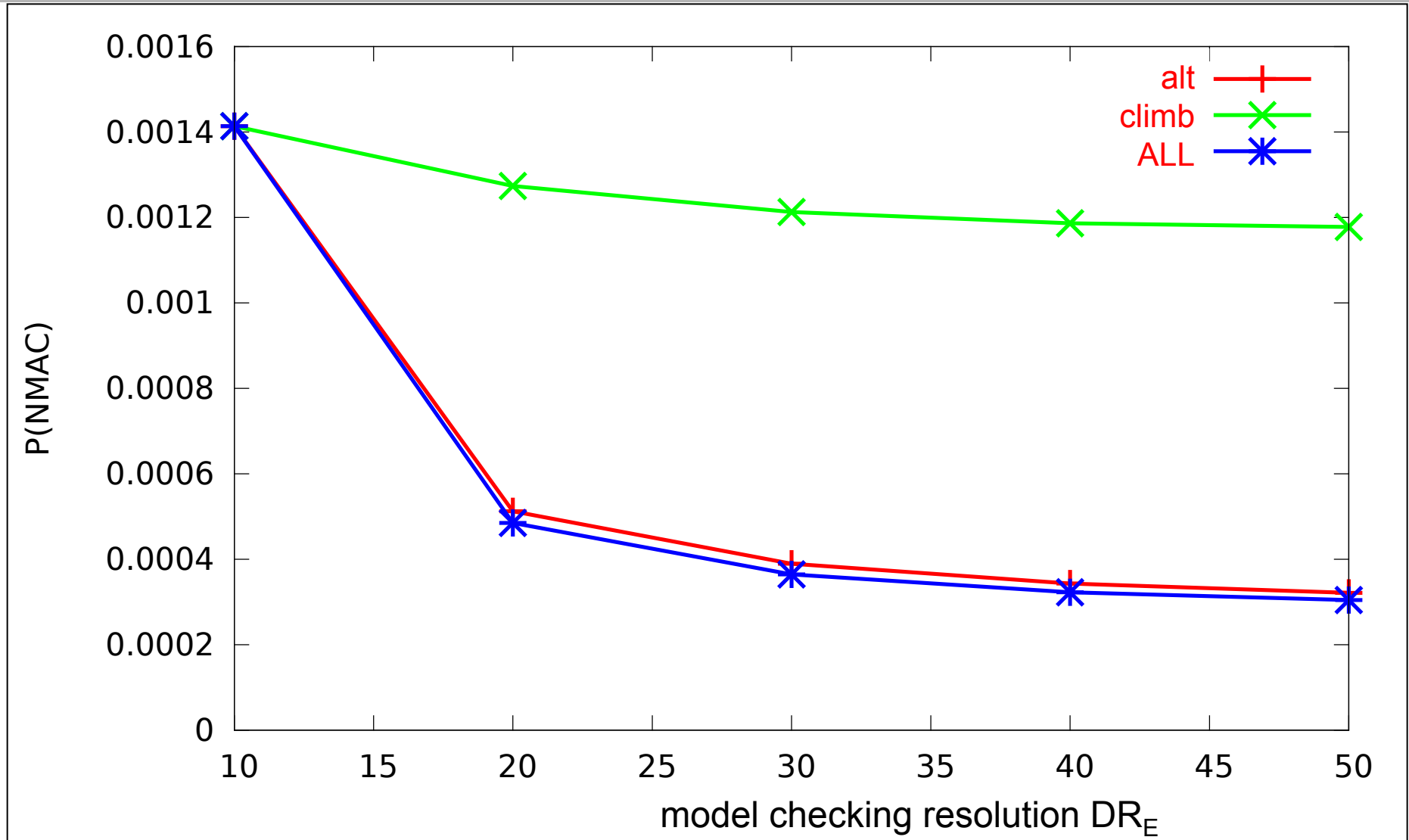
DR_E : model discretization resolution to model check look-up table



effects of resolution DR_E on model checking

- we compute $P(NMAC)$ of the baseline look up table deployed in models that are obtained through discretization with varying resolutions DR_E (dh_{own} , dh_{int} , h_{rel})
 - **ALL** varies climb rates and relative altitude in DR_E : (20, 20, 20), (30, 30, 30), ...
 - **climb** varies climb rates only in DR_E : (20, 20, 10), (30, 30, 10), ...
 - **alt** varies relative altitude only in DR_E : (10, 10, 20), (10, 10, 30), ...

effects of resolution DR_E on model checking



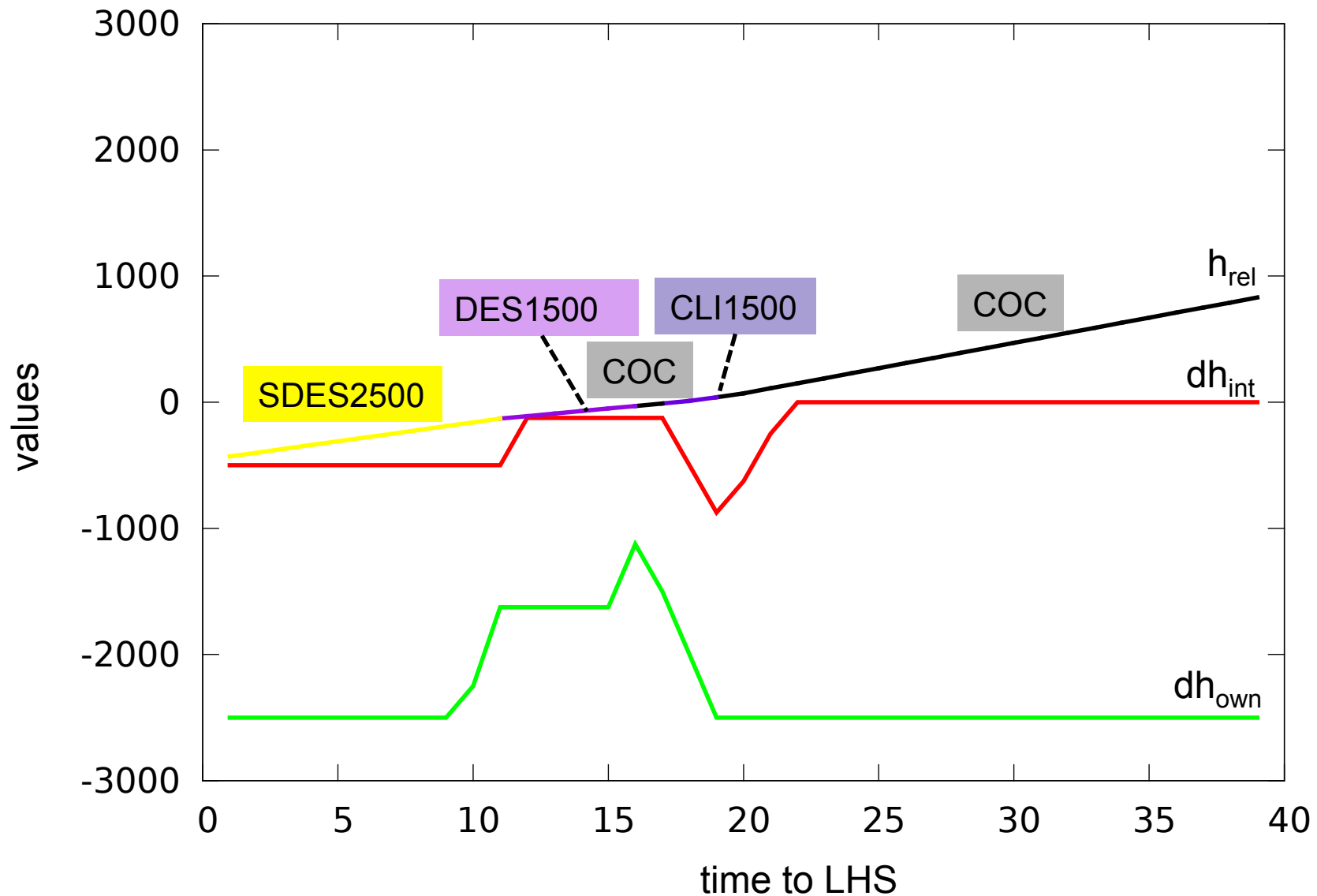
- P(NMAC) decreases with higher evaluation resolutions
- relative altitude discretization is indicative

probabilistic model checking $DR_E = (50, 50, 100)$

allows precise *automated analysis* of probabilistic properties expressed in a formal logic such as *PCTL*; generates *encounters* that exhibit property-related behaviors

- what is the probability of **NMAC**? ($P=?[F \text{ NMAC}]$) 2.5×10^{-4}
- what if pilot responds immediately?
($P=?(F \text{ NMAC} \mid Ga_{\text{prev}} = s_{RA})$) 2.3×10^{-8}
- what is the probability of a **split advisory**? 1.8×10^{-3}
 $P=?[F (!COC \wedge P=1[X \text{ COC}] \wedge P>0 [F !COC])]$
- split advisories are harder to directly take into account during look up table generation because they require to record history

split advisory encounter

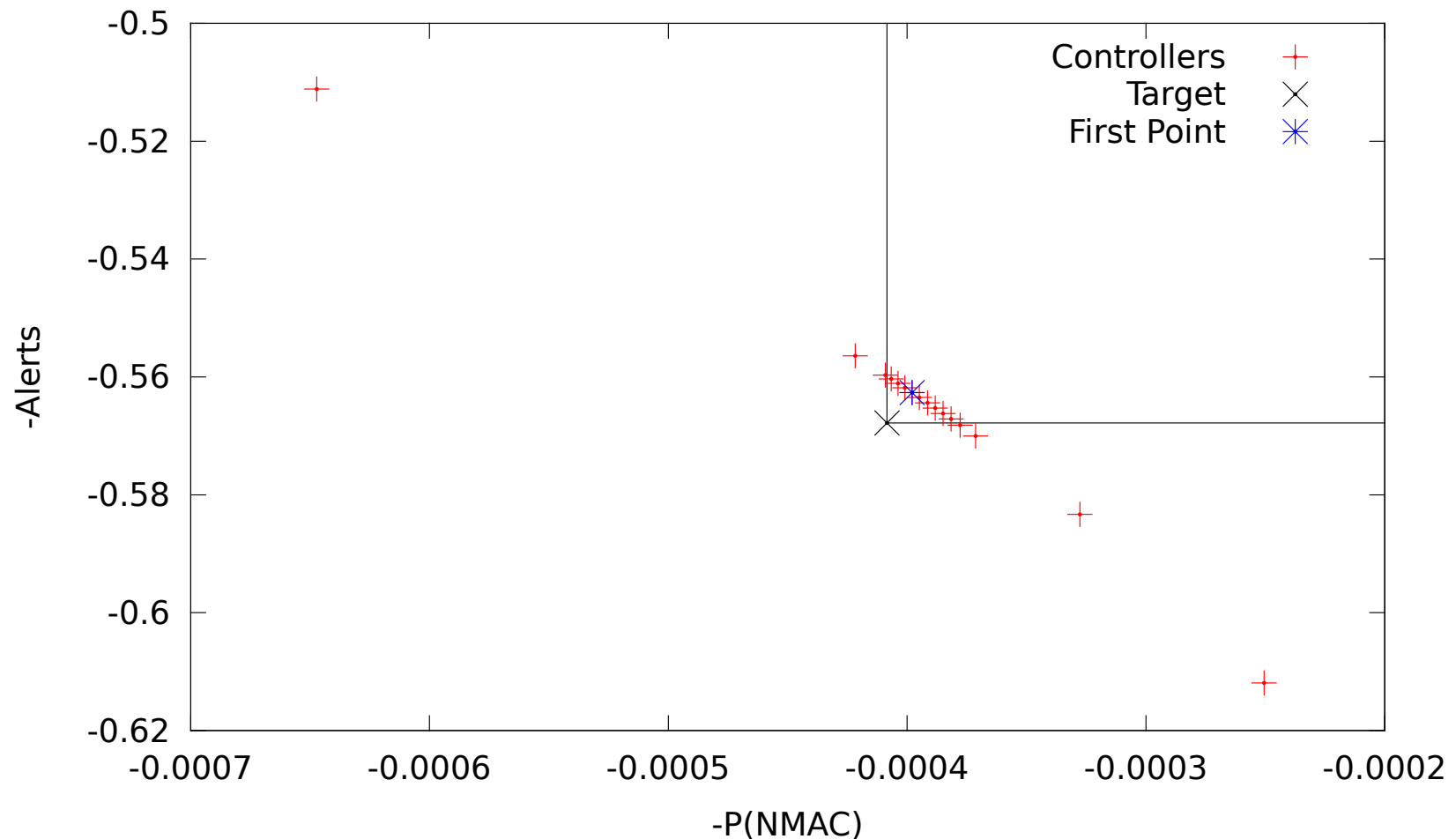


(reward for COC + cost of alert) < cost of reversal ("sneaky" reversals)

synthesis / design

weights vs. performance

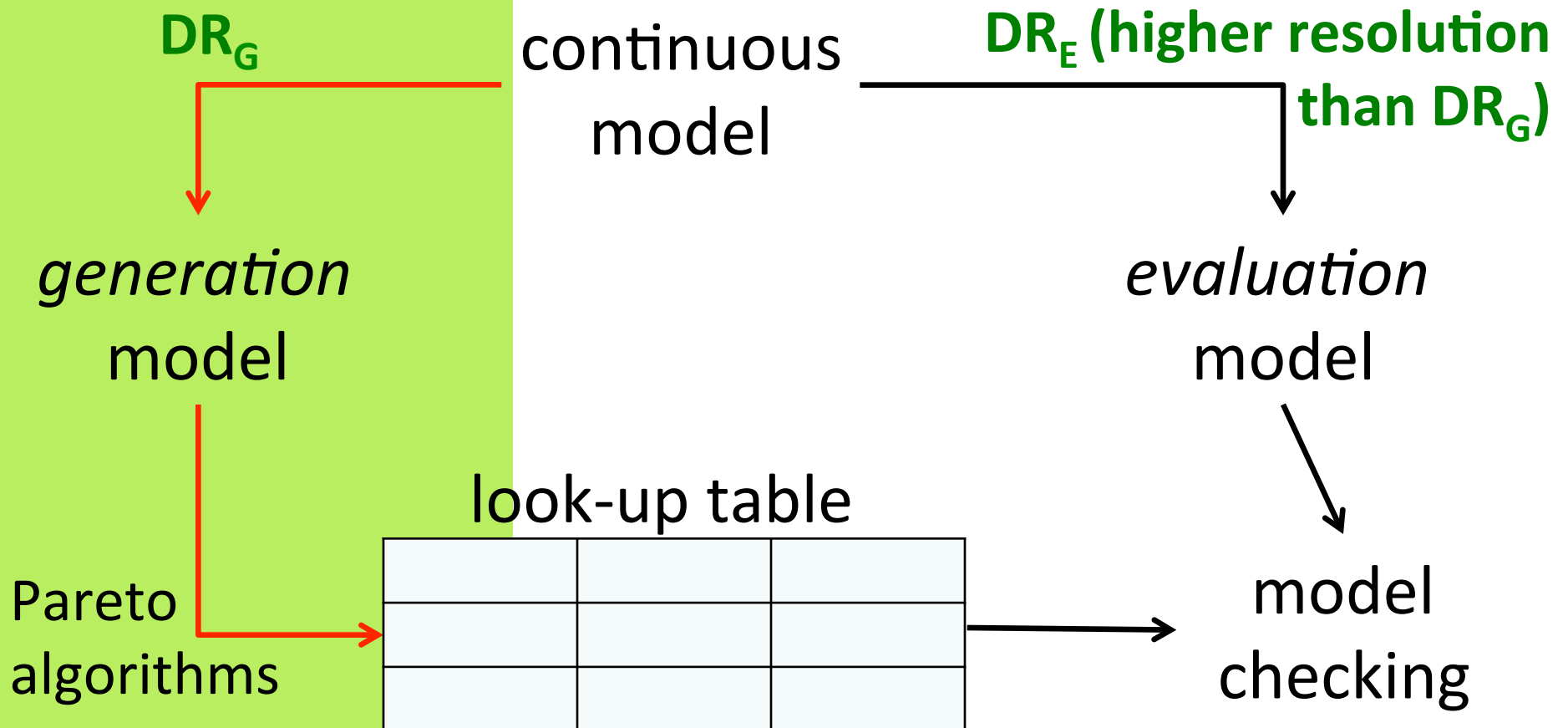
- tune look-up tables based on minimum acceptable performance
 - deterministic look-up tables based on weights form a convex Pareto front; we implement algorithms that approximate it *above* target performance



discretization resolution (dh_{own} , dh_{int} , h_{rel})

DR_G : model discretization resolution for look-up table generation;
baseline [KC 2011] resolution is ($dh_{own}=10$, $dh_{int}=10$, $h_{rel}=10$)

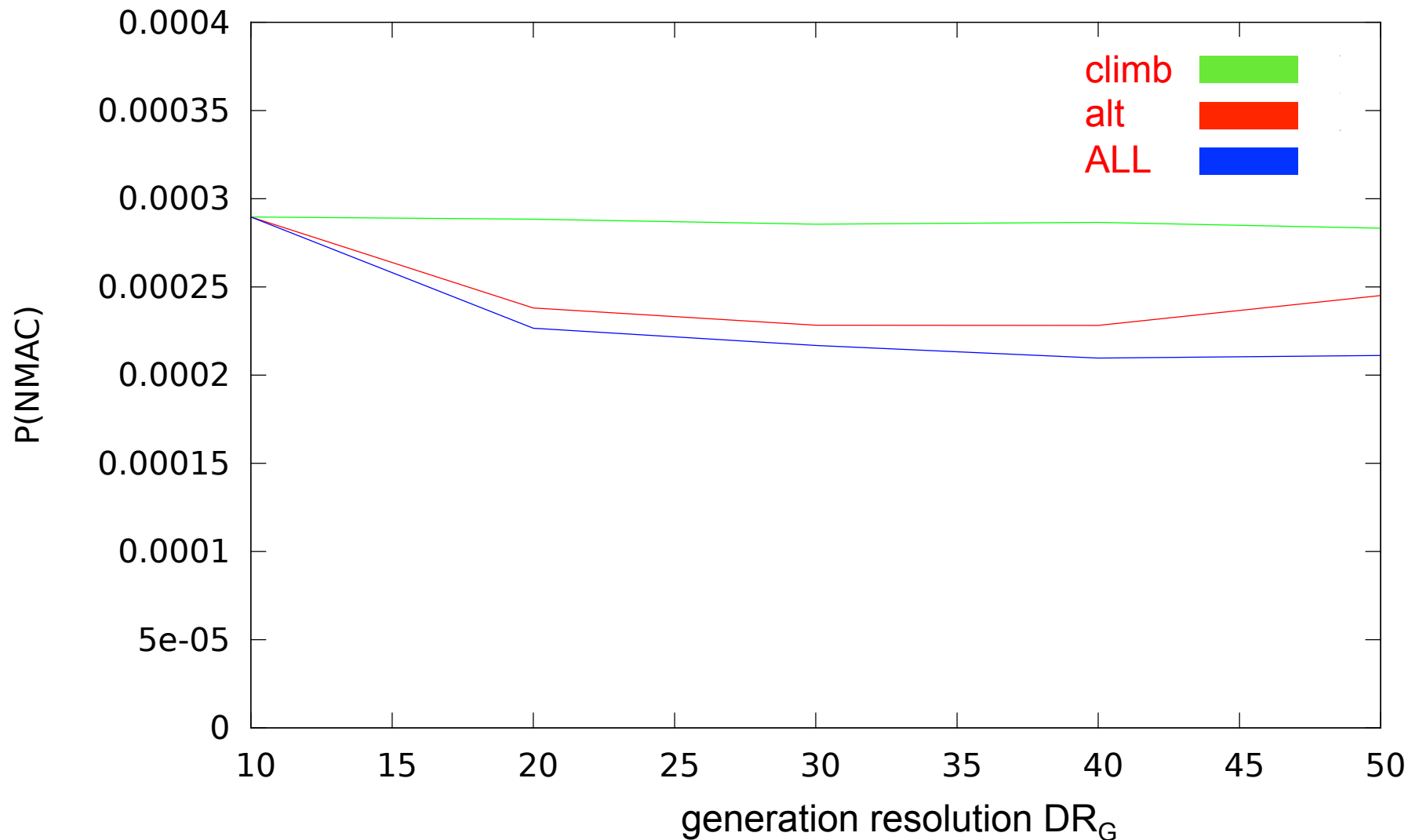
DR_E : model discretization resolution to model check look-up table



synthesizing better tables in higher DR_G

- question: what is the effect of resolution discretization DR_G on look-up table synthesis?
- experiment set up:
 - evaluate baseline ($dh_{own}=10$, $dh_{int}=10$, $h_{rel}=10$) [KC 2011] in new DR_G
 - use result as target for synthesis
- how we vary resolutions DR_G (dh_{own} , dh_{int} , h_{rel})
 - **ALL** varies climb rates and relative altitude in DR_G : (20, 20, 20), (30, 30, 30), ...
 - **climb** varies climb rates only in DR_G : (20, 20, 10), (30, 30, 10), ...
 - **alt** varies relative altitude only in DR_G : (10, 10, 20), (10, 10, 30), ...
 - *note that **alt** results in the smallest look up tables – in terms of numbers of states – for each value increase*
- compare the synthesized look-up tables in $DR_E = (50, 50, 100)$

table synthesis at different resolutions



recommendation: (10, 10, 30), or (n, n, 3*n)

verification achievements

- we could not use off-the-shelf tools, so we built [VeriCA](#) toolset
 - our tools support models written in Java
 - we customized verification and synthesis algorithms for ACAS X needs
- we analyzed ACAS X version that we reproduced based on:

Kochenderfer, M. J., and Chryssanthacopoulos, J. P. Robust airborne collision avoidance through dynamic programming. Project Report ATC-371, Massachusetts Institute of Technology, Lincoln Laboratory, 2011.
- [ETAPS 2014 EASST best paper award](#)
 - Christian von Essen, Dimitra Giannakopoulou: *Analyzing the Next Generation Airborne Collision Avoidance System*, TACAS 2014.
- [FAA / NASA Ames Interagency Agreement](#) for ACAS X and VeriCA
 - thus able to apply our subsequent work on the actual ACAS X code

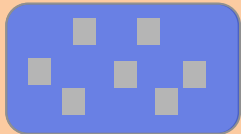
model quality

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

We **defined** Conformance Relations that compare MDP model to flight data

t



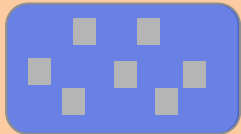
state estimate at time t
on look-up table

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

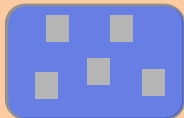
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t



state estimate at time t
on look-up table

t+1sec

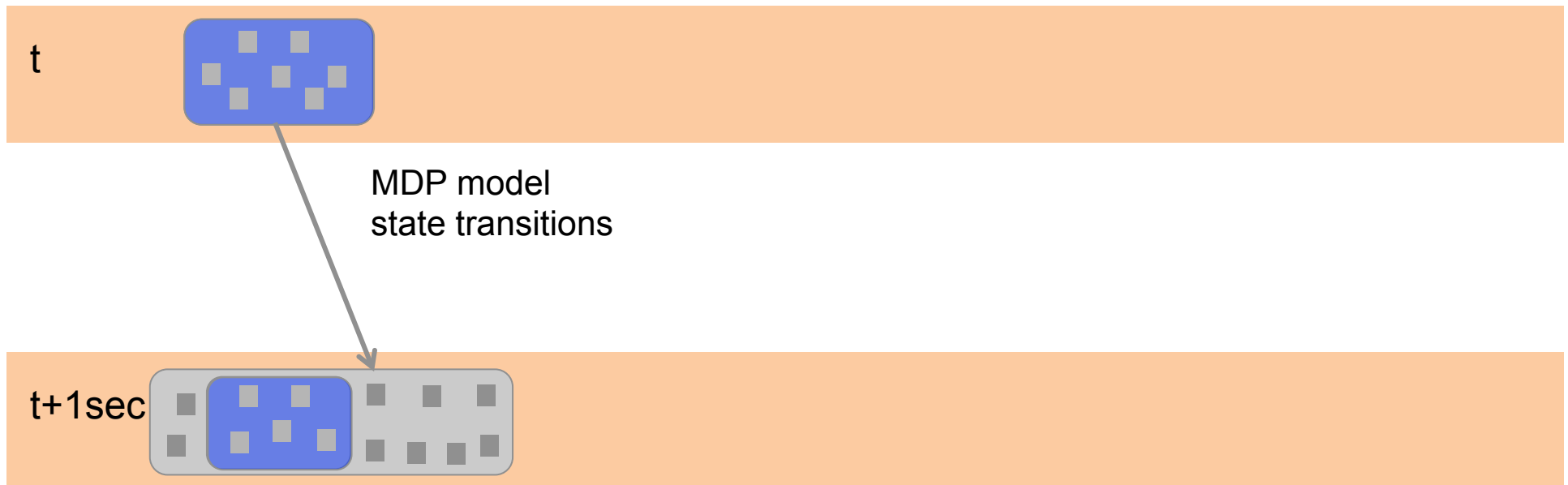


state estimate at time t+1
on look-up table

comparing model to real world

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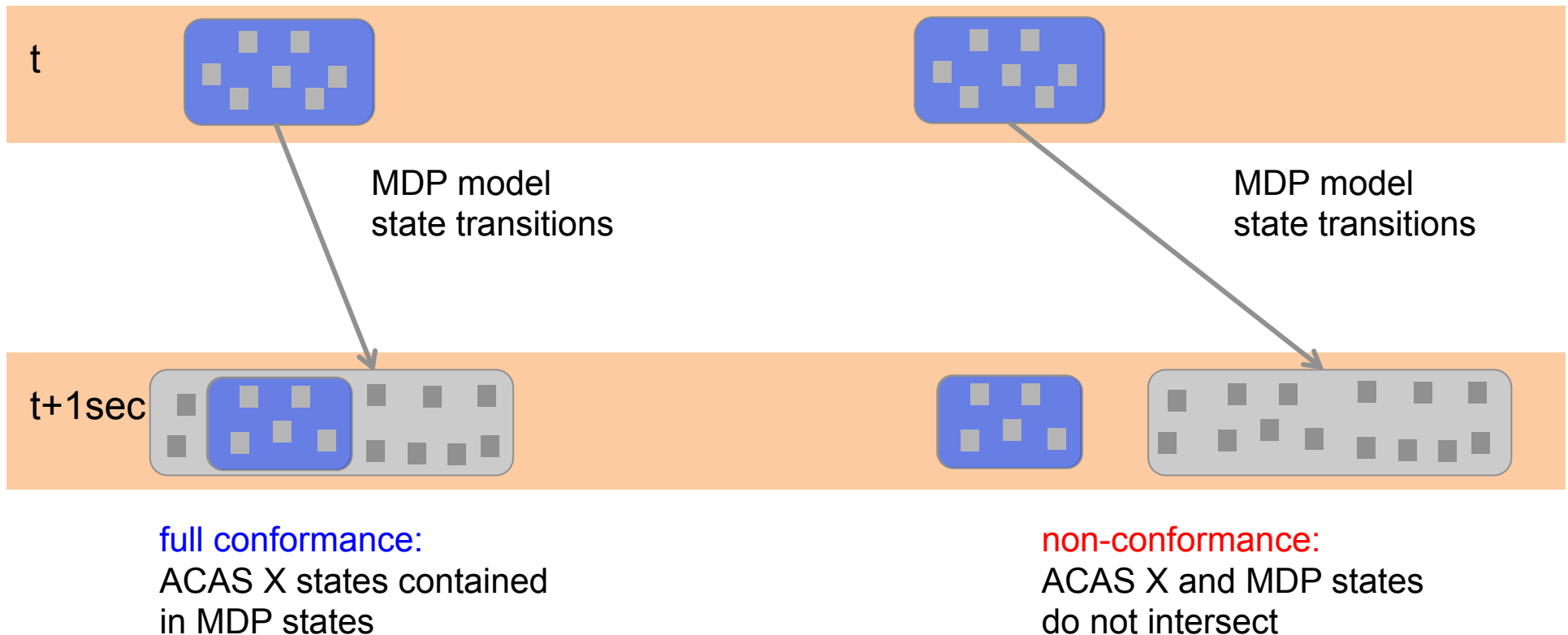
full conformance:

ACAS X states contained
in MDP states

comparing model to real world

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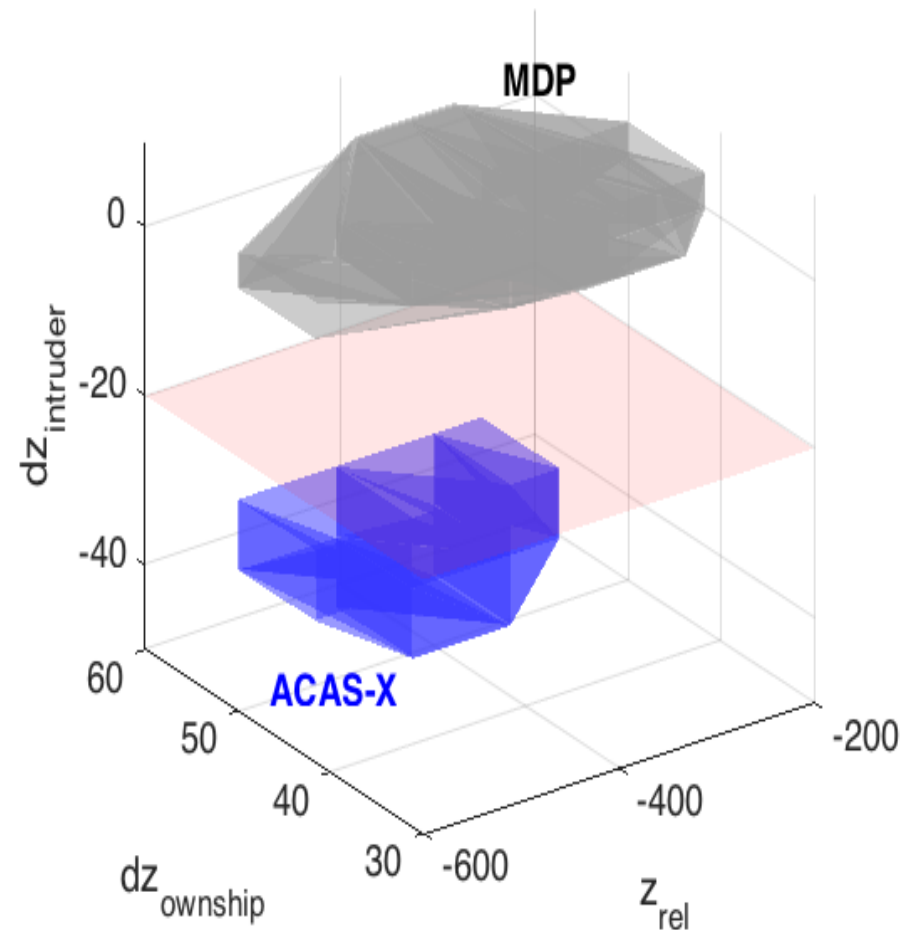


data generation and analysis

Data generation: Non-conforming encounters are rare in test data of the ACAS X distribution. We used a reinforcement learning framework to target generation of non-conforming simulated encounters.

Data Analysis: The intruder climb rate has been identified as a common factor for divergence across the data. *Further analysis is needed.*

Open question: Does non-conformance imply potential violation of safety requirements?



applicability

self-driving cars

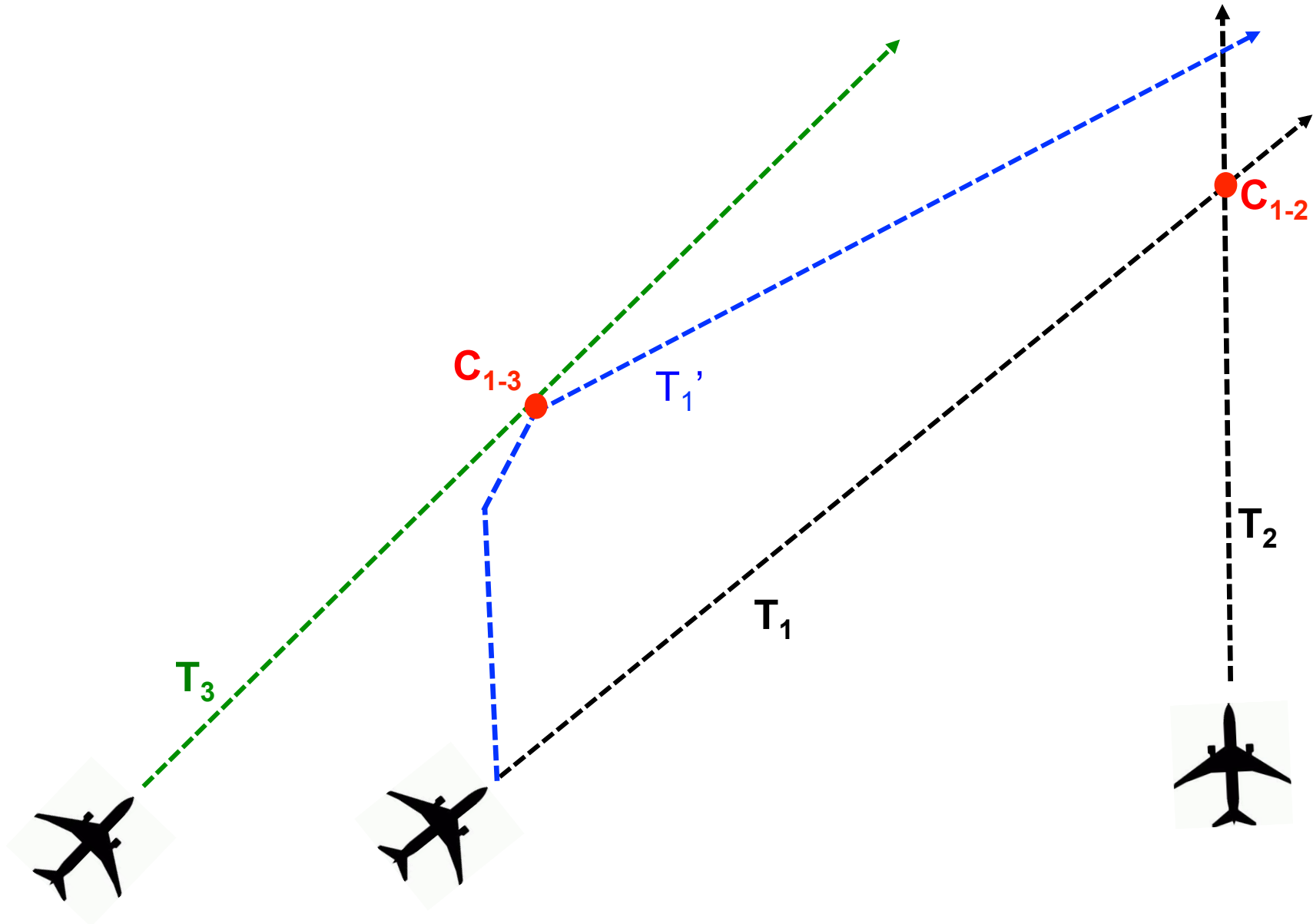


V&V of autonomy

- formulation of requirements is harder – autonomy-specific?
 - optimization, adaptive and learning algorithms
 - example: loss of separation, ACAS X

generate strategic secondary aircraft

no picked resolution is allowed to cause a more imminent secondary conflict



V&V of autonomy

- formulation of requirements is harder – autonomy-specific?
 - optimization, adaptive and learning algorithms
 - example: separation assurance, ACAS X
- need for advanced testing infrastructures
 - test case generation for stress-testing and requirements coverage
 - examples: ACAS X, separation assurance, autonomous vehicles
- V&V at runtime, including requirements
 - ACAS X (error prediction with statistical learning)
 - separation assurance
- trust
 - extensive verification
 - explanation of decision-making algorithms

Collaborators / Publications

1. Dimitra Giannakopoulou, David H. Bushnell, Johann Schumann, Heinz Erzberger, Karen Heere: **Formal testing for separation assurance**. Ann. Math. Artif. Intell. 63(1): 5-30 (2011)
2. Dimitra Giannakopoulou, Falk Howar, Malte Isberner, Todd Lauderdale, Zvonimir Rakamaric, Vishwanath Raman: **Taming test inputs for separation assurance**. ASE 2014.
3. Marko Dimjasevic, Dimitra Giannakopoulou: **Test-Case Generation for Runtime Analysis and Vice Versa: Verification of Aircraft Separation Assurance**. ISSTA2015.
4. Christian von Essen, Dimitra Giannakopoulou: **Probabilistic verification and synthesis of the next generation airborne collision avoidance system**. STTT 18(2): 227-243 (2016). Extended version of TACAS 2014 paper awarded ETAPS 2014 EASST best paper.
5. Dimitra Giannakopoulou, Dennis Guck, Johann Schumann: **Exploring Model Quality for ACAS X**. FM 2016.